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# MAXIMAL ORDER OF MULTIPOINT ITERATIONS USING n EVALUATIONS\*

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### ABSTRACT

This paper deals with multipoint iterations without memory for the solution of the nonlinear scalar equation  $f^{(m)}(x)=0, \ m\geq 0. \ \text{Let } p_n(m) \text{ be the maximal order of iterations which use n evaluations of the function or its derivatives per step. We prove the Kung and Traub conjecture <math display="block">p_n(0)=2^{n-1} \text{ for Hermitian information. We show } p_n(m+1)\geq p_n(m) \text{ and conjecture } p_n(m)\equiv 2^{n-1}. \ \text{The problem of the maximal order is connected with Birkhoff interpolation. Under a certain assumption we prove that the Polya conditions are necessary for maximal order.}$ 

### INTRODUCTION

We consider the problem of solving the nonlinear scalar equation  $f^{(m)}(x) = 0$  where m is a nonnegative integer. We solve this problem by multipoint iterations without memory which use n evaluations of the function or its derivatives per step. For fixed n we seek an iteration of maximal order of convergence. This problem is connected with Birkhoff interpolation and can be expressed in terms of the incidence matrix  $E_n^k = (e_{ij})$  where  $e_{ij} = 1$  if  $f^{(j)}(z_i)$  is computed and

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 $e_{ij} = 0$  otherwise;  $z_i \neq z_j$ , and  $\sum_{i=1}^{k} \sum_{j=0}^{\infty} e_{ij} = n$ . (Note that the problem of Birkhoff interpolation has been open for 70 years, see Sharma [72].)

Let  $p_n(m)$  be the maximal order of multipoint iterations. For m=0, Kung and Traub showed that  $p_n(0)\geq 2^{n-1}$ . We show that  $p_n(m+1)\geq p_n(m)$  and conjecture  $p_n(m)=2^{n-1}$ . For m=0 we prove the Kung and Traub conjecture for Hermitian information, i.e., if  $f^{(j)}(z_j)$  is computed, then  $f^{(0)}(z_j),\ldots,f^{(j-1)}(z_i)$  are also computed. Under a certain assumption we prove that the Pólya conditions are necessary for the maximal order, i.e., the total number of  $f,f',\ldots,f^{(j)}$  evaluations has to be at least j+1,  $j=0,1,\ldots,n-1$ . We show also that  $p_n(0)\leq n(n+1)^{n-1}$ . Some special incidence matrices  $E^k$  are considered and maximal orders of iterations based on  $E^k$  are discussed.

### 2. THE n-EVALUATION PROBLEM

We consider the problem of solving the nonlinear scalar equation

$$(2.1)$$
  $f^{(m)}(x) = 0$ 

where  $f \colon D_F \subset \mathbb{C} \to \mathbb{C}$ ,  $\mathbb{C}$  denotes the one dimensional complex space and m is a nonnegative integer. We assume that there exists a simple zero  $\alpha$  of  $f^{(m)}f^{(m)}(\alpha) = 0 \neq f^{(m+1)}(\alpha)$ , and that f is analytic in a neighborhood of  $\alpha$ . Let  $\Im$  denote a class of such functions.

We solve (2.1) by stationary iteration and assume that  $\mathbf{x}_1$  is a sufficiently close approximation to  $\alpha$ . To get the next approximation  $\mathbf{x}_2$  to  $\alpha$  we need some information on f. We assume that this information  $\mathfrak{N}=\mathfrak{N}(\mathbf{x}_1;\mathbf{f})$  is given by some

values of the function and its derivatives at the points z defined as follows. Let

$$z_1 : f^{(j_1)}(z_1), \dots, f^{(j_{\mu_1})}(z_1), \dots, f^{(j_{\mu_1})}(z_1), \dots, f^{(j_{\mu_k})}(z_k)$$
 $\vdots$ 
 $z_k : f^{(j_1)}(z_k), \dots, f^{(j_{\mu_k})}(z_k)$ 

denote points and numbers of derivatives which are computed where nonnegative integers  $\{j_{i,i}^{\,\,i}\}$  satisfy the relations

$$j_{\mu}^{i} < j_{\mu+1}^{i}$$
 for i=1,2,...,k and  $\mu$ =1,2,..., $\mu_{i}^{-1}$ ,
$$\mu_{1} + \mu_{2} + ... + \mu_{k} = n.$$

Furthermore,

(2.2) 
$$z_{1} = x_{1}$$
  $z_{i+1} = z_{i+1}(z_{1}, ..., z_{i}, f^{(j_{1})}(z_{1}), ..., f^{(j_{\mu_{1}})}(z_{1}), ..., f^{(j_{\mu_{1}})}(z$ 

This means that every  $z_{i+1}$  is the function of the previous information computed at  $z_1, \ldots, z_i$  and the next approximation  $x_2 = z_{k+1}$  depends on n evaluations. Sometimes we shall use the notation  $z_i = z_i(x_i)$  or  $z_i = z_i(x_i, f)$  to stress the dependence on  $x_i$  and f.

To simplify further notations we define an incidence  $\underline{\text{matrix}} \ E_n^k = (e_{ij})$  of the information  $\mathfrak{N}$ ,  $i=1,2,\ldots,k$  and  $j=0,1,\ldots$ , as follows. Let

(2.3) 
$$e_{ij} = \begin{cases} 1 & \text{if we compute } f^{(j)}(z_i) \\ 0 & \text{if we do not compute } f^{(j)}(z_i), \end{cases}$$

where

where 
$$\infty$$
(2.4)  $\sum_{j=0}^{\infty} e_{j} > 0$  for  $i = 2,3,...,k$ ,

(2.5) 
$$|E_n^k| = \sum_{i=1}^k \sum_{j=0}^{\infty} e_{ij} = n$$
, (thus  $k \le n+1$ ).

The condition (2.4) means that at every point  $z_i$ ,  $i \ge 2$ , we compute at least one derivative. (We consider f to be the zeroth derivative  $f^{(0)}$ .) However we do not, at this point, insist on any information being computed at  $z_1 = x_1$ . We show in Lemma 3.2 that  $f^{(m)}$  must be evaluated at  $x_1$ . The condition (2.5) means that we use exactly n evaluations. Let

(2.6) 
$$e_n^k = \{(i,j): e_{ij} = 1, i = 1,2,...,k; j = 0,1,...\}$$

Hence the information  $\mathfrak R$  can then be defined in terms of the incidence matrix  $E_n^k$  as follows:

(2.7) 
$$\mathfrak{N} = \mathfrak{N}(x_1; f) = \{f^{(j)}(z_i) : (i,j) \in e_n^k\}.$$

The concept of an incidence matrix is used in Birkhoff interpolation, see Sharma [72]. We shall show some connections between the n evaluation problem and Birkhoff interpolation.

Having the information  $\mathfrak N$  we define the next approximation  $x_2$ ,  $x_2 = z_{k+1}$ , as  $x_2 = \varphi(x_1; \Re(x_1; f))$  where  $\varphi$  is a given function.

We call  $\phi$  an iteration function if for every  $f\in \S$ , with  $f^{(m)}(\gamma) = 0$  there exists  $\delta > 0$  such that for any  $x_1$ ,  $|x_1-\alpha| \le \delta$ , the sequence

(2.8a) 
$$x_{d+1} = \phi(x_d; \Re(x_d; f)), d = 1,2,...$$

is well-defined and

(2.8b) 
$$\lim_{d\to\infty} x_d = \alpha$$
,

(2.8c) 
$$\alpha = \omega(\alpha, \Re(\alpha; f)).$$

Such iterations are called k-point iteration without memory since they use exactly n new evaluations at k distinct points. If k > 1 they are called <u>multipoint iterations</u> (see Traub [61], [64], and Kung and Traub [74]). Let  $\Phi$  be a class of iterations  $\Phi$  with  $k \geq 1$ .

Since these iterations are stationary and without memory it is sufficient to define how  $\mathbf{x}_2$  is generated from  $\mathbf{x}_1$  and to measure the goodness of  $\phi$  by examining some properties of  $\mathbf{x}_2$  -  $\alpha$  as  $\mathbf{x}_1$  tends to  $\alpha$ .

We want to find an iteration for which  $\mathbf{x}_2$  approximates  $\alpha$  as closely as possible, i.e., we seek an iteration with the maximal order. In a previous paper (Wozniakowski [75]) we proved that if a set of iterations  $\Phi$  is not empty then the maximal order of iteration is equal to the order of information. This gives us a powerful technique for proving maximal order. Let us briefly recall what we mean by orders of iteration and information.

We shall say  $\{\widetilde{f}(\cdot; x_1)\}$  is equal to f with respect to  $\mathfrak{N}$  (briefly denoted by  $\widetilde{f}$   $\overline{\mathfrak{N}}$  f) iff

(i) f, 
$$f(\cdot; x_1) \in \mathcal{I}$$
,

(ii) 
$$\tilde{f}^{(m)}(\tilde{\alpha}; x_1) = 0$$
 and  $f^{(m)}(\alpha) = 0$  where  $\tilde{\alpha} = \tilde{\alpha}(x_1)$  and  $\lim_{x_1 \to \alpha} \tilde{\alpha}(x_1) = \alpha$ ,

(iii) 
$$\lim_{x_1 \to \alpha} \tilde{f}^{(j)}(\alpha; x_1) = g^{(j)}(\alpha)$$
 where  $g(\alpha) = 0$  and  $g \in \mathcal{R}$ ,  $j = 0,1,...$ 

(iv) 
$$\mathfrak{N}(x_1; \tilde{f}) = \mathfrak{N}(x_1; f)$$
, i.e.,  $\tilde{f}^{(j)}(z_i; x_1) = f^{(j)}(z_i)$   
for  $(i,j) \in e_n^k$ .

The first three conditions mean that  $\tilde{f}(x; x_1)$  is sufficiently regular with respect to x and tends to a function g,  $g \in \mathfrak{J}$ , as  $x_1$  tends to  $\alpha$ . The condition (iv) means that  $\tilde{f}$  and f have the same information  $\mathfrak{N}$  at the point  $x_1$ . Therefore any iteration  $\phi$  will produce the same approximation  $x_2$  for  $\tilde{f}$  and f,  $\phi(x_1; \mathfrak{N}(x_1; \tilde{f})) \equiv \phi(x_1; \mathfrak{N}(x_1; f))$ . Since we cannot recognize  $\tilde{f}$  from f using information (2.7), we should approximate not only the zero  $\alpha$  of f, but at the same time, the zero  $\tilde{\alpha}$  of  $\tilde{f}$ . This leads us to the following definitions of orders of iteration and information.

Let A be a set defined by

$$A=\left\{q\geq 1; \ \forall f\in \mathfrak{J}, f^{(m)}(\alpha)=0, \ \forall \widetilde{f}\ \overline{\widetilde{g}}\ f, \ \limsup_{x_{1}\to \alpha}\frac{\left|x_{2}-\widetilde{\alpha}\right|}{\left|x_{1}-\alpha\right|^{q-\varepsilon}}=0, \ \forall \varepsilon>0\right\}$$

A number  $p = p(\phi)$  is called an order of the iteration  $\phi$  iff

(2.9) 
$$p(\phi) = \begin{cases} 0 & \text{if A is empty,} \\ \sup A & \text{otherwise.} \end{cases}$$

Using this convention  $p(\phi)$  always exists; however the only interesting cases are for  $A \neq \emptyset$ . Furthermore, let

$$B = \{q \ge 1; \forall f \in \mathfrak{I}, f^{(m)}(\alpha) = 0, \forall \tilde{f} = f, \lim \sup_{x_1 \to \alpha} \frac{|\alpha - \tilde{\alpha}|}{|x_1 - \alpha|^{q - \epsilon}} = 0, \forall \epsilon > 0\}.$$

A number  $p = p(\mathfrak{N})$  (sometimes denoted  $p = p(E_n^k)$ ) is called <u>an</u> order of the information  $\mathfrak{N}$  if

(2.10) 
$$p(\mathfrak{N}) = \begin{cases} 0 & \text{if B is empty,} \\ \sup B & \text{otherwise.} \end{cases}$$

We know that if  $\Phi \neq \emptyset$  then

(2.11) 
$$\sup_{\varphi \in \Phi} p(\varphi) = p(\mathfrak{N})$$

and  $p(\mathfrak{N}) = p(I_{\mathfrak{N}})$  where  $I_{\mathfrak{N}}$  is a generalized interpolatory method. (See Wozniakowski [75].)

We are now in a position to define the n-evaluation problem (see Kung and Traub [73] and [74]). For fixed n and m we wish to find a number k = k(n,m), points  $z_i = z_i(x_1)$  for  $i=2,3,\ldots,k$ , an incidence matrix  $E_n^k$ ,  $|E_n^k|=n$ , and an iteration  $\phi$  which uses  $E_n^k$  (see (2.8)) such that  $p(\phi)$  is maximal. Due to (2.11) this is equivalent to maximizing the order of information  $\Re$ , i.e., to find  $E^{*k}$  such that

(2.12) 
$$p_n(m) = \sup_{E_n} p(E_n^k),$$

(2.13) 
$$p(E*_{n}^{k}) = p_{n}(m)$$
.

We recall the <u>Kung and Traub conjecture</u> for m = 0 (Kung and Traub [74]):

$$(2.14)$$
  $p_n(0) = 2^{n-1}$ .

They showed two different matrices  $E_n^k$ ,  $n \ge 2$ , for which the order of iteration is equal to  $2^{n-1}$  (see Section 3), so we know that

$$(2.15) \quad p_n(0) \ge 2^{n-1}.$$

We now show a relationship among the  $p_n(m)$  for different m.

### Lemma 2.1

Let  $\phi=\phi(\mathfrak{N})$  be an iteration of order p for the problem  $f^{(m)}(x)=0$  which uses n evaluations per step. Then there exists an iteration  $\phi^*=\phi^*(\mathfrak{N}^*)$  for the problem  $f^{(m+1)}(x)=0$  which also uses n evaluations and has the same order p.

### Proof

Let  $E_n^k = (e_{ij}^k)$  be the incidence matrix of  $\mathfrak{R}$  and  $E_n^{*k} = (e_{ij}^k)$  be defined by

$$e_{ij}^{*} = \begin{cases} 1 & \text{if } e_{i,j-1} = 1 \\ 0 & \text{otherwise.} \end{cases}$$

Let  $\mathfrak{A}^*$  be information with the incidence matrix  $\mathbf{E}_n^{*k}$  based on the points  $\mathbf{z}_i = \mathbf{z}_i(\mathbf{x}_1)$ ,  $i = 2, \dots, k$ , from  $\mathfrak{A}$ . For any  $\mathbf{f}_1$  from  $\mathfrak{B}$ ,  $\mathbf{f}_1^{(m+1)}(\alpha) = 0 \neq \mathbf{f}_1^{(m+2)}(\alpha)$ , define

$$f(x) = f'_1(x).$$

Thus,  $f \in \mathcal{R}$ ,  $f^{(m)}(\alpha) = 0 \neq f^{(m+1)}(\alpha)$ , and  $f^{(j)}(x) \equiv f_1^{(j+1)}(x)$ . Hence

$$\mathfrak{N}^*(x_1; f_1) = \mathfrak{N}(x_1; f).$$

Let us define φ by

$$\phi^*(x_1; \, \mathfrak{N}^*(x_1; \, f_1)) = \phi(x_1, \, \mathfrak{N}(x_1; \, f)).$$

Since  $f_1$  is arbitrary it easily follows that  $p(\phi^*) = p(\phi)$ . From Lemma 2.1 and (2.15) we immediately get

### Corollary 2.2

$$p_n^{(m)} \ge p_n^{(m-1)} \ge 2^{n-1}$$
 for any  $m \ge 1$ .

Although Corollary 2.2 states that  $\mathbf{p}_{\mathbf{n}}(\mathbf{m})$  is at least  $\mathbf{p}_{\mathbf{n}}(\mathbf{m-1})$  we propose

Conjecture 2.3
$$p_{n}(m) = 2^{n-1} \quad \forall m \ge 0, n \ge 1.$$

### 3. EXISTENCE OF ITERATIONS

Recall that  $\Phi$  is a class of iterations defined by (2.8). In this section we show what we have to assume on the information  ${\mathfrak R}$  to be sure that  $\Phi$  is not empty. We shall prove that  $\Phi = \emptyset$  if any of the following three conditions hold:

- (1) If  $z_i(x_1)$  does not converge to  $\alpha$ .
- (2) If we do not compute  $f^{(m)}(x_1)$ , i.e.,  $e_{1m} = 0$ .
- If n = 1 under the assumption on sufficiently regularity of  $\phi$  as a function of  $x_1$ .

We prove this in the following Lemmas.

### Lemma 3.1

Let  $\phi$  be an iteration which uses the information  $\mathfrak{N}_{\bullet}$ Then for any  $f \in \mathfrak{F}$ ,  $f^{(m)}(\alpha) = 0$ ,

$$\lim_{x_1 \to \alpha} z_i(x_1; f) = \alpha \quad \text{for } i = 1, 2, \dots, k+1.$$

### Proof

Suppose on the contrary that there exist  $f \in \mathfrak{I}$ ,  $f(\alpha) = 0$ , an index i,  $2 \le i \le k$ , a number  $\epsilon > 0$  and a sequence  $\{x_i\}$ such that

$$\lim_{j\to\infty} x_j = \alpha \quad \text{and} \quad |z_i(x_j) - \alpha| \ge \epsilon \quad \text{for } j \ge j_0.$$

Let  $J=\{x\colon |x-\alpha|<\varepsilon\}$ . Define  $f_1\colon J\to\mathbb{C}$  such that  $f_1(x)=f(x)$  for  $x\in J$ . Since  $f_1\in\mathfrak{J}$  there exists  $\delta_1>0$ such that any  $x_1$ ,  $|x_1-\alpha| \le \delta_1$  is a good initial approximation. Setting  $x_1 = x_j$ , for large j, where  $|x_j - \alpha| \le \delta_1$ , we get  $z_i(x_j) \not\in J$  and  $\mathfrak{N}(x_1; f_1)$  is not well defined which contradicts (2.8a).

### Lemma 3.2

Let  $\mathfrak{N}$  be any information with the incidence matrix  $E_n^k$ . If  $\phi \neq \emptyset$  then  $e_{1m} = 1$ , (i.e. we have to compute  $f^{(m)}(x_1)$ ).

Compare Theorem 4.1 in Kung and Traub [73] which proves this result for m = 0.

### Proof

Let  $\varphi \in \Phi$  and suppose on the contrary that  $e_{1m} = 0$ . Let f be any function from  $\Im$ ,  $f^{(m)}(\alpha) = 0$ . Let  $x_1$  be sufficiently close approximations to  $\alpha$ ,  $x_1 \neq \alpha$ . From (2.2) we get  $\delta = \min_{2 \leq i \leq k} |z_i(x_1) - x_1| > 0$ .

Define
$$f_{1}(x) = \begin{cases} f(x) - \frac{f^{(m)}(x_{1})}{m!}(x-x_{1})^{m} & \text{for } |x-x_{1}| < \delta \\ f(x) & \text{otherwise} \end{cases}$$

Note that  $f_1 \in \mathfrak{J}$ ,  $f_1^{(m)}(x_1) = 0$ , and

$$f_1^{(j)}(x_1) = f^{(j)}(x_1)$$
 for  $j \neq m$   
 $f_1^{(j)}(z_i) = f^{(j)}(z_i)$  for any j and  $i = 2,...,k$ .

Since we do not compute  $f^{(m)}(x_1)$  then

$$\Re(x_1; f_1) = \Re(x_1; f).$$

But  $x_1$  is the zero of  $f_1$  and due to (2.8c) it follows

$$x_2 = \varphi(x_1; \Re(x_1; f)) = \varphi(x_1; \Re(x_1; f_1)) = x_1.$$

Thus,  $x_d = x_1$  and  $\lim_{d} x_d \neq \alpha$  which contradicts (2.8b).

An iteration function  $\varphi$  can be treated as a function of x,  $\varphi(x) = \varphi(x; \Re(x; f))$  for x close to  $\varphi$ . We shall prove that if  $\varphi$  is sufficiently regular then the number of evaluations n has to be at least two.

### Lemma 3.3

If an iteration  $\phi$  is a sufficiently smooth function of x then  $n \ge 2$ .

### Proof

It is enough to prove Lemma 3.3 for the real case. Assume on the contrary that n=1. From Lemma 3.2 it follows that this unique piece of information is given by  $f^{(m)}(x_1)$ . Let

$$\varphi(x; f^{(m)}(x)) = x + g(x, f^{(m)}(x)).$$

From (2.8b) it follows

$$g(\alpha, 0) = 0$$
  $\forall \alpha \text{ such that } f^{(m)}(\alpha) = 0, f \in \mathfrak{I}$ 

From this and the regularity of  $\phi$  we can express g(x, y)

$$g(x, y) = y^k h(x, y)$$

for an integer  $k \ge 1$  where  $h(x, 0) \ne 0$  and h(x) = h(x, f(x)) is a continuous function for x close to  $\alpha$ .

Let  $h(\alpha) \neq 0$  and for simplicity we assume that  $h(\alpha) > 0$ . (If  $h(\alpha) < 0$  then the proof is analogous.) Let  $f \in \mathfrak{J}$  be a polynomial of degree m+1 and  $f^{(m+1)}(x) \equiv 1$ ,  $f(\alpha) = 0$ . There exists  $\delta = \delta(f) > 0$  such that for any  $x_1$ ,  $\begin{vmatrix} x_1 - \alpha \end{vmatrix} \leq \delta$  the sequence  $x_{d+1} = \phi(x_d)$ ,  $f^{(m)}(x_d) = x_d + \begin{bmatrix} f^{(m)}(x_d) \end{bmatrix} h(x_d)$  is well defined for any d and converges to  $\alpha$  (see (2.8)). For

 $e_d = x_d - \alpha we get$ 

(3.2) 
$$e_{d+1} = [1 + e_d^{k-1} h(x_d)] e_d$$

If  $x_1$  is close but different from  $\alpha$  then  $e_d \neq 0$  for any d. Since  $\lim_{d \to 0} e_d = 0$  then for any  $d_1$  there exists  $d \geq d_1$  such that  $|e_{d+1}|^d < |e_d|$ , i.e.

(3.3) 
$$|1 + e_d^{k-1} h(x_d)| < 1$$
.

We consider two cases.

Case I. Let k be odd. Then for large d we have

$$e_d^{k-1} h(x_d) \approx e_d^{k-1} h(\alpha) > 0$$

which contradicts (3.3).

Case II. Let k be even. We prove that h does not change sign for  $x \in [\alpha - \delta, \alpha + \delta]$ . If so, then by the continuity of h there exists  $x^*$  such that  $h(x^*) = 0$  and  $0 < |x^* - \alpha| < \delta$ . Setting  $x_1 = x^*$  we get  $x_d = x^*$  which contradicts (3.3). Thus  $h(x) \ge h_0 > 0$  for  $|x - \alpha| \le \delta$ . Define  $f_1 \colon [\alpha - \delta, \alpha + \delta] \to \Re$  such that  $f_1(x) = f(x)$ . Since  $f_1$  also belongs to  $\Im$ ,  $f_1^{(m)}(\alpha) = 0$ , there exists  $\delta_1 > 0$  such that  $x_{d+1} = \varphi(x_d; \Re(x_d; f_1))$  is well defined whenever  $|x_1 - \alpha| \le \delta_1$ . Let  $x_1 > \alpha$ . Keeping in mind that  $\Re(x_d; f_1) \equiv \Re(x_d; f)$ , from (3.2) we get

$$e_{d+1} \ge (1 + e_d^{k-1}h_0)e_d \ge (1 + e_1^{k-1}h_0)^d e_1.$$

Hence, there exists an index d such that  $e_{d+1} > \delta$ , and since  $f_1(x_{d+1})$  is not defined we get a contradiction with (2.8a).

### 4. HERMITIAN INFORMATION

In this section we deal with a special case of the nevaluation problem when the information  ${\mathfrak N}$  is hermitian.

### Definition 4.1

 $\mathfrak{N}$  is called <u>hermitian information</u> if the incidence matrix  $\textbf{E}_n^k$  (which is now called hermitian) satisfies

$$e_{ij} = 1 \Rightarrow e_{i0} = e_{i1} = \dots = e_{i,j-1} = 1 \quad \forall (i,j) \in e_n^k$$

This means that if  $f^{(j)}(z_i)$  is computed then  $f^{(0)}(z_i),...,$   $f^{(j-1)}(z_i)$  are also computed.

Let  $s_i$  denote the number of evaluations at  $z_i$ , i.e.,  $e_{i,s_i}^{-1} = 1$  and  $e_{i,s_i}^{-1} = 0$ . Then

(4.1) 
$$s_1 + s_2 + ... + s_k = n$$
 where  $s_i \ge 1$  for  $i = 1, 2, ..., k$ .

For given n and k we want to find s<sub>i</sub> and z<sub>i</sub>, i = 1,2,...,k, to maximize the order of information. Let  $p_n(m,H)$  be the maximal order of hermitian information. Note that  $p_n(m) \ge p_n(m,H)$ .

First we shall discuss a property of hermitian informations for the problem f(x) = 0, i.e., m = 0.

# Theorem 4.1 (m = 0)

The order  $p(E_{n}^{k})$  of the hermitian information  $\mathfrak R$  with the incidence matrix  $E_{n}^{k}$  satisfies

(4.2) 
$$p(E_n^k) \le s_1 \prod_{i=2}^k (s_i+1).$$

## Proof

It is easy to verify that if  $\tilde{f}$   $\bar{\tilde{g}}$  f then

(4.3) 
$$\tilde{f}(x; x_1) = f(x) + G(x; x_1) \prod_{i=1}^{k} (x-z_i)^{s_i}$$

for an analytic function G. Since  $\tilde{f}'(\alpha; x_1)$  tends to  $g'(\alpha) \neq 0$  then setting  $x = \alpha$  in (4.3) we get

$$(4.4) \quad (\alpha - \widetilde{\alpha}) = \frac{G(\alpha; x_1)}{g'(\alpha)} (1 + o(1)) \prod_{i=1}^{k} (\alpha - z_i)^{s_i}.$$

Define q, by

$$\frac{\frac{\alpha^{-z}i}{q_i-\varepsilon} \to 0 \quad \text{and} \quad \frac{\frac{\alpha^{-z}i}{q_i+\varepsilon} \to +\infty, \quad \forall \varepsilon > 0}{e_1}$$

where  $e_1 \equiv x_1 - \alpha$ . Since  $z_i = z_i(x_1)$  tends to  $\alpha$  (see Lemma 3.1) then  $q_i$  exists and  $q_i \ge 0$  for i = 1, 2, ..., k. Note that  $q_1 = 1$ .

Let  $p_1 = q_1 = 1$  and

(4.5) 
$$p_{j+1} = \sum_{i=1}^{j} q_{i} s_{i}, \quad j = 1, 2, ..., k.$$

From (4.4) we get

$$(4.6) \quad \frac{\alpha - \tilde{\alpha}}{\underset{e_{1}}{p_{k+1} - \epsilon}} = \frac{G(\alpha; x_{1})}{g'(\alpha)} (1 + o(1)) \prod_{i=1}^{k} \left\{ \frac{\alpha - z_{i}}{\underset{e_{1}}{q_{i} - \delta}} \right\}^{s_{i}} \rightarrow 0, \forall \epsilon > 0,$$

where  $\delta = \epsilon/n$ . For  $G(\alpha; x_1) \equiv \text{const} \neq 0$  we get

(4.7) 
$$\frac{\alpha - \tilde{\alpha}}{p_{k+1} + \epsilon} \to \infty , \forall \epsilon > 0.$$

Now we shall prove that there exists a function f such that

(4.8) 
$$q_i \le p_i$$
 for  $i = 1, 2, ..., k$ .

Let f be any function such that  $f \in \mathfrak{J}$ ,  $f(\alpha) = 0$  and  $f^{(j)}(\alpha) \neq 0$  for  $j = 1, 2, \ldots$ . Since  $p_1 = q_1$ , the condition (4.8) holds for i = 1. Assume by induction that this holds for  $i \leq j$ . Suppose by the contrary that

$$q_{j+1} > p_{j+1} = \sum_{i=1}^{j} q_i s_i$$

Define

$$r = \sum_{i=1}^{s} i$$

<u>Case I.</u> Let r = 1. This means that j = 1,  $s_1 = 1$  and  $z_2 = z_2(x_1, f(x_1))$  approximates q with order greater than  $p_2 = 1$ .

Define

(4.9) 
$$h(x_1, f(x_1)) = \frac{x_1 - f(x_1) - z_2}{z_2 - x_1} + 1.$$

It is easy to verify that

$$h(x_1, f(x_1)) = f'(\alpha)(1 + o(1)).$$

Case II. Let r>1 and  $\widetilde{f}$  be the Hermite interpolatory polynomial of degree less than r defined by

$$\tilde{f}^{(1)}(z_i) = f^{(1)}(z_i), \quad i = 1,2,...,j; \ 1 = 0,1,...,s_{i-1}.$$

Let  $\tilde{\alpha}$  be the nearest zero of  $\tilde{f}$  to  $z_1 = x_1$ . Then

$$(4.10) \quad \frac{\tilde{\alpha} - \alpha}{j} s_{i} f'(\alpha) = \frac{f^{(r)}(\alpha)}{r!} (1 + o(1)).$$

$$\prod_{i=1}^{n} (\alpha - z_{i})^{s_{i}}$$

Note that  $\widetilde{\alpha}$  is a function of  $x_1$  and information  $\mathfrak{N}$   $(x_1; f) = \{f^{(1)}(z_i): i = 1, 2, \dots, j; 1 = 0, 1, \dots, s_i^{-1}\}$  Recall that  $z_{j+1} = z_{j+1}(x_1, \mathfrak{N}(x_1; f))$  and  $z_{j+1} - \alpha = o(e_1)$ . Define

(4.11) 
$$h(x_1, \Re(x_1; f)) = \frac{\tilde{\alpha} - z_{j+1}}{\int_{i=1}^{j} (z_{j+1} - z_i)^{s_i}} \tilde{f}'(z_{j+1}).$$

Thus h is the lefthand side of (4.10) where  $\alpha$  is replaced by  $z_{j+1}$ . Since  $z_{j+1}$  is a better approximation to  $\alpha$  than  $\tilde{\alpha}$ , it is straightforward to verify that

(4.12) 
$$h(x_1, \Re(x_1; f)) = \frac{f^{(r)}(\alpha)}{r!}(1 + o(1)).$$

This means that in both cases using r evaluations of the function and its derivatives given by  $\mathfrak N$  we can approximate the rth normalized derivative. We prove that this is impossible.

Note that h (see (4.9) or (4.11)) is a continuous function of  $x_1$  at  $x_1 = \alpha$  and

(4.13) 
$$h(\alpha, \mathfrak{N}(\alpha; f)) = \frac{f^{(r)}(\alpha)}{r!}$$
.

Let  $f_1(x) = f(x) + (x-\alpha)^r$  and let us apply h to the function  $f_1$ . Thus

$$h(\alpha, \mathfrak{N}(\alpha; f)) = h(\alpha, \mathfrak{N}(\alpha; f_1)) = \frac{f^{(r)}(\alpha)}{r!} + 1$$

which contradicts (4.13).

Hence  $q_{j+1} \le p_{j+1}$  which proves (4.8). Keeping in mind  $p(E_n^k) = p_{k+1}$  and using (4.5), (4.8) we get

$$p(E_{n}^{k}) = \sum_{i=1}^{k} q_{i} s_{i} \leq \sum_{i=1}^{k} p_{i} s_{i} = \sum_{i=1}^{k-1} p_{i} s_{i} + p_{k} s_{k} \leq (1+s_{k}) \sum_{i=1}^{k} p_{i} s_{i}$$

$$\leq s_1 \prod_{i=2}^{k} (s_i+1)$$

which proves Theorem 4.1.

We want to show that a bound in (4.2) is sharp, i.e., there exist points  $z_2, \ldots, z_k$  such that the order of information is equal to  $s_1 = \sum_{i=1}^{k} (s_i + 1)$ .

Let w<sub> $\mu$ </sub>,  $\mu$  = 1,2,...,k, be the Hermite interpolatory polynomial of degree less than r<sub> $\mu$ </sub> = s<sub>1</sub> + s<sub>2</sub> + ... + s<sub> $\mu$ </sub> defined by

(4.14) 
$$w_{\mu}^{(j)}(z_i) = f^{(j)}(z_i), i = 1,2,...,\mu; j = 0,1,...,s_i-1.$$

Let  $\alpha_{\mu}$  be the nearest zero of  $w_{\mu}$  to  $z_1 = x_1$ . (If  $s_1 = 1$  then  $\alpha_1 = x_1 - \beta f(x_1)$  for any nonzero constant  $\beta$ .)

Define  $z_{\mu+1}$  as a point such that

(4.15) 
$$z_{\mu+1} = \alpha_{\mu} + O(e_1^{\beta_{\mu}}), \beta_{\mu} \ge s_1 \prod_{i=2}^{\mu} (s_i+1).$$

From (4.14) it follows

$$(4.16) \quad \alpha_{\mu} - \alpha = \begin{cases} (\beta f'(\alpha) - 1)(\alpha - z_{1}) + o(\alpha - z_{1}) & r_{\mu} = 1 \\ \frac{(r_{\mu})}{f'(\alpha)} & \prod_{i=1}^{\mu} (\alpha - z_{i})^{s_{i}} + o(\prod_{i=1}^{\mu} (\alpha - z_{i})^{s_{i}}) & \text{if } r_{\mu} > 1. \end{cases}$$

From (4.15) we get

(4.17) 
$$z_{\mu+1} - \alpha = O(e_1^{q_{\mu+1}}), q_{\mu+1} = s_1 \prod_{i=2}^{\mu} (s_i+1),$$

which proves that the order of information  $\mathfrak{R}$  based on the points  $z_{\mu+1}$  from (4.15) is equal to  $s_1 \prod_{i=2}^{k} (s_i+1)$ .

An iteration which uses this information  $\Re$  and has the maximal order can be defined as follows.

For 
$$\mu = 1, 2, ..., k$$

- (i) construct  $\psi$  from (4.14) using a divided-difference algorithm,
- (ii) apply Newton iteration to the equation w(x) = 0 setting

$$y_0 = z_{\mu}$$
 $y_{i+1} = y_i - w'_{\mu}(y_i)^{-1}w_{\mu}(y_i), i = 0,1,...,i_0-1,$ 
 $z_{\mu+1} = y_i$ 

where

$$(4.18)$$
  $i_0 = \lceil \log_2(s_{u+1} + 1) \rceil$ .

(If  $s_1 = 1$  then  $z_2 = x_1 - \beta f(x_1)$ .) Then (4.15) holds and

(4.19) 
$$z_{k+1} - \alpha = O(e_1^{q_{k+1}}), q_{k+1} = s_1 \prod_{i=2}^{k} (s_i+1).$$

Furthermore if  $\beta_{\mu} > q_{\mu+1}$  in (4.15) then we can specify the constant which appears in the "0" notation in (4.19). Note that  $\beta_{\mu} > q_{\mu+1}$  if we redefine  $i_0$  in (4.18) as the smallest integer such that  $i_0 > \log_2(s_{\mu+1} + 1)$ .

### Lemma 4.2

Let  $\phi$  be the iteration defined as above,  $z_{k+1}=\phi(x_1,\,\Re(x_1;\,\,f))$  . If  $\beta_{\mu}>q_{\mu+1}$  for  $\mu$  = 1,2,...,k then

(4.20) 
$$\lim_{\substack{x_1 \to \alpha \\ x_1 \to \alpha}} \frac{z_{k+1}(x_1) - \alpha}{(x_1 - \alpha)} = C_{k+1}$$

where

$$C_{\mu+1} = M_r \prod_{j=1}^{\mu-1} M_r j^{(s_{j+2}+1)...(s_{\mu}+1)}$$
 for  $\mu = 1, 2, ..., k$ 

and

$$M_{i} = \begin{cases} (-1)^{i} \frac{f^{(i)}(\alpha)}{i!f'(\alpha)} & \text{if } i > 1 \\ -\beta f'(\alpha) + 1 & \text{if } i = 1. \end{cases}$$

Ιf

$$(4.21) \quad \underline{K}^{i-1} \leq \left| \frac{f^{(i)}(\alpha)}{i!f'(\alpha)} \right| \leq \overline{K}^{i-1} \quad \text{for } i = r_1, r_2, \dots, r_k$$

then

$$(4.22) \quad c \cdot \underline{K}^{q_{k+1}-1} \leq \lim_{x_1 \to \alpha} \left| \frac{z_{k+1}(x_1) - \alpha}{q_{k+1}} \right| \leq \overline{K}^{q_{k+1}-1} \cdot c$$

where

$$c = \begin{cases} 1 & \text{if } r_1 > 1 \\ |M_1| & \text{s}_2(s_3+1)\dots(s_k+1) \\ |M_1| & \text{if } r_1 = 1 \text{ and } k \ge 2 \end{cases}$$

$$|M_1| & \text{if } r_1 = 1 \text{ and } k = 1$$

Note that the righthand side of (4.21) follows from the analyticity of f.

Proof

Let  $C_i = \lim_{x_1 \to \alpha} (z_1 - \alpha) / (x_1 - \alpha)^{q_i}$ . Note that  $C_1 = 1$ . From (4.15), (4.16) and since  $\beta_{\mu} > q_{\mu+1}$  we get

$$z_{\mu+1} - \alpha = \alpha_{\mu} - \alpha + z_{\mu+1} - \alpha_{\mu} = M_{r_{\mu} i=1} (z_{i} - \alpha)^{s_{i}} + o(e_{1}^{q_{\mu+1}}).$$

Thus

(4.23) 
$$C_{\mu+1} = M_r \prod_{i=1}^{\mu} C_i^s$$

Since  $C_1 = 1$  we get after some tedious calculations

$$C_{\mu+1} = M_{r} \prod_{j=1}^{\mu-1} M_{r_{j}}^{s_{i+1}} (s_{i+2}^{+1}) \dots (s_{\mu}^{+1})$$

which proves the first part of Lemma 4.2.  $q_i - 1 \qquad q_i - 1 \qquad q_$ This is true for i = 1 since  $C_1 = q_1 = 1$ . From (4.23) and (4.21) we have

$$|c_{\mu+1}| \le \bar{k}^{\mu-1} + s_1(q_1-1) + \dots + s_{\mu}(q_{\mu}-1) = \bar{k}^{\mu+1}$$

and similarly we get a lower bound.

Let  $r_1 = 1$ . Assume by induction that  $c_i \underbrace{K}^{q_i} \leq |C_i| \leq K^{q_i} c_i$  where  $c_1 = 1$ ,  $c_2 = |M_1|$  and  $c_{i} = |M_{1}|^{s_{2}(s_{3}+1)...(s_{i-1}+1)}$  for  $i \ge 3$ . This is true for i = 1 and 2 since  $C_1 = q_1 = q_2 = 1$  and  $C_2 = M_{r_1}$ . Then

$$|c_{\mu+1}| \leq \bar{K}^{q_{\mu+1}-1} |M_1| \leq 2^{+s_2 s_3 + s_4 s_2 (s_3 + 1) \dots s_{\mu} s_2 (s_3 + 1) \dots (s_{\mu-1} + 1)} = \bar{K}^{q_{\mu+1}-1} c_{\mu+1}$$

and similarly we get a lower bound. Hence (4.22) holds which

completes the proof.

Lemma 4.2 in the case  $r_1>1$  states that the asymptotic constant  $C_{k+1}$  depends exponentially on the order  $q_{k+1}$ . This property makes an analysis of the complexity of iteration easier (Traub and Wozniakowski will analyze it in a future paper).

We are now in a position to answer the following question. For given n and k,  $k \le n$ , find nonnegative integers  $s_1, s_2, \ldots, s_k$  to maximize the order of information

 $p_k = \max_{\substack{s_1 + \dots + s_k = n}} s_1 \prod_{i=2}^k (s_i+1)$ . Using a standard technique it is easy to verify that

$$(4.24) \quad \left(n + (k-1) \left\lceil \frac{n-1}{k} \right\rceil \right) \left(1 + \left\lceil \frac{n-1}{k} \right\rceil \right)^{k-1} \le p_k \le \left(\frac{n+k-1}{k}\right)^k < 2^{n-1}$$

for  $k \le n-2$  and  $p_k = 2^{n-1}$  for k = n-1 or n. If k is a divisor of n-1 then the optimal  $s_i$  are given by

$$s_1 = 1 + \frac{n-1}{k}$$
 and  $s_i = \frac{n-1}{k}$  for  $i = 2, ..., k$ .

For k = n the optimal  $s_i \equiv 1$ . Furthermore from Theorem 7.1 in Kung and Traub [74] it follows that there are exactly two cases which maximize the order of information,

$$k = n-1$$
,  $s_1 = 2$ ,  $s_i = 1$  for  $i = 2,...,n$ ,  $p_{n-1} = 2^{n-1}$   
 $k = n$ ,  $s_i = 1$  for  $i = 1,...,n$ ,  $p_n = 2^{n-1}$ .

The first case means that we use f and f' at the first point and f at the other points. The second case states that we use n function evaluations. From Theorem 4.1 and (4.24) we get

### Corollary 4.3

The Kung and Traub conjecture holds for hermitian information  $(p_n(0,H) = 2^{n-1})$ .

The next part of this section deals with the general problem  $f^{(m)}(x) = 0$ ,  $m \ge 1$ . It seems to us that hermitian information is not always relevant for that problem especially for large m. Note that we have to compute  $f^{(m)}(x_1)$  and if the information is hermitian then we have to assume  $n \ge m+1$ . On the other hand if we use  $f^{(m)}(z_1), \ldots, f^{(m)}(z_n)$  (which is nonhermitian) then the order of information is  $2^{m-1}$ . However it is interesting to know the optimal order of information for special hermitian cases, e.g., f, f' at  $z_1$  followed by n-1 function evaluation at the other points for the problem f'(x) = 0, (see Lemma 4.5).

Recall that  $p_n(m,H)$  denotes the maximal order of hermitian information. In general we do not know  $p_n(m,H)$ . We only show some bounds on it.

$$\frac{\text{Lemma 4.4}}{p_n(m,H)} \leq 2^{n-1}.$$

<u>Proof</u>

If  $\tilde{f} = f$  then

(4.25) 
$$\tilde{f}^{(m)}(x) - f^{(m)}(x) = [G(x) | \prod_{i=1}^{k} (x-z_i)^{s_i}]^{(m)}$$

for an analytic function G. Let  $G(x) = \frac{1}{m!}(x-\alpha)^m$ . Since  $\tilde{f}^{(m+1)}(\alpha)$  tends to  $g^{(m+1)}(\alpha) \neq 0$  as  $x_1$  tends to  $\alpha$  then setting  $x = \alpha$  in (4.25) we have

$$\tilde{\alpha} - \alpha = c(\alpha, x_1) \prod_{i=1}^{k} (\alpha - z_i)^{s_i}$$

where  $c(\alpha, x_1)$  tends to a nonzero limit (see (4.4)).

The proof of Lemma 4.4 may now be obtained analogously to the proof of Theorem 4.1.

### Lemma 4.5

Let  $n \ge m+1 \ge 2$ . Then

$$p_n(m,H) \ge c q(m)^{n-1}$$

where

$$c = c(m) = \frac{2}{(1+2m+\sqrt{t})}, q(m) = \left(\frac{1+\sqrt{t}}{2}\right)^{m}$$

and t = 1 + 4m.

### Proof

Define  $s_1 = m+1$  and  $s_i = m$  for i = 2,...,k. Let  $z_2 = x_1 + \beta$   $f^{(m)}(x_1)$  for  $\beta \neq 0$  and let  $z_1, \mu \geq 3$ , be the nearest zero to  $z_{\mu-1}$  of the polynomial  $w_{\mu}^{(m)}$  where

$$w_{\mu}^{(j)}(z_{i}) = f^{(j)}(z_{i}), \quad i = 1,2,...,\mu-1;$$
  
 $j = 0,1,...,m-1,$   
 $w_{\mu}^{(m)}(z_{1}) = f^{(m)}(z_{1})$ 

and w is of degree  $\leq (\mu-1)m_{\bullet}$  It is straightforward to verify that

$$z_{\mu} - \alpha = O((x_1 - \alpha)^{q_{\mu}})$$

where  $q_1 = q_2 = 1$  and for  $\mu \ge 3$ ,

$$q_{\mu} = m(q_1 + \dots + q_{\mu-2}) + q_1 = q_{\mu-1} + mq_{\mu-2}.$$

It is easy to verify that

$$q_{k+1} \geq c \left(\frac{1+\sqrt{t}}{2}\right)^{k+1}$$
 where  $c = c(m) = 2/(1+2m+\sqrt{t})$ .

For a given n let  $k = \lfloor (n-1)/m \rfloor = \frac{n-1}{m} + \theta$  where  $-1 < \theta \le 0$ . The total number of evaluation is equal to  $km + 1 \le n$ . Hence  $p_n(m, H) \ge p_{km+1}(m, H) \ge q_{k+1} \ge$ 

Lemma 4.4 and 4.5 state that  $p_n(m,H)$  as a function of n is exponentially bounded from below and above. However  $\lim_{m\to\infty} q(m) = 1.$ 

### 5. GENERAL INFORMATION, m = 0

We deal with the n-evaluation problem for m = 0. For small n it is possible to verify the Kung and Traub conjecture and to characterize the information sets for all iterations which have maximal order.

For n=1 the unique piece of information is given by  $f(x_1)$ . Since  $\widetilde{f}(x)=f(x)+(x-x_1)$  has the same information as f then  $p_1(0)=1$ . This means that for any  $y=y(x_1,f(x_1))$  the distance  $\alpha$ -y can be at most of first order in  $\alpha$ - $x_1$ . However y is not, in general, an iteration function, see Lemma 3.3. Note also that for any m,  $p_1(m)=1$ .

For n = 2, Kung and Traub [73] proved that the maximal order of iteration equals two under a certain assumption on the iterations considered. Using our technique we find the order of information for any  $\mathbb R$  with n = 2. Note that if  $\mathbb R$  is hermitian information then  $p(\mathbb R) \leq 2$ , by Corollary 4.3. Thus it suffices to consider the non-hermitian case. Let us first consider one-point iterations, i.e., k = 1 and  $\mathbb R = \{f(x_1), f^{(j)}(x_1)\}$  for  $j \geq 2$ . Then  $f(x) = f(x) + (x-x_1)$  and  $p(\mathbb R) = 1$ . Let us pass to two-point iterations, i.e., k = 2 and  $\mathbb R = \{f(x_1), f^{(j)}(z_2)\}$  where  $j \geq 1$  and

 $z_2 = z_2(x_1, f(x_1))$ . If  $j \ge 2$  then  $\tilde{f}(x) = f(x) + (x-x_1)$  and  $p(\mathfrak{N}) = 1$ . Let j = 1. Then  $\tilde{f}(x) = f(x) + (x-x_1)(x-2z_2+x_1)$ . From this we get

$$\tilde{\alpha}$$
- $\alpha \cong (\alpha - x_1)(\alpha - y), \quad y = 2z_2 - x_1.$ 

Since  $y = y(x_1, f(x_1))$  then  $\alpha$ -y can be at most of first order in  $(\alpha-x_1)$ . Hence  $p(\mathfrak{N}) \leq 2$  and  $p(\mathfrak{N}) = 2$  if, for instance,  $z_2 = x_1 + \beta f(x_1)$ , for any constant  $\beta \neq 0$ .

It is easy to verify that, in addition,  $p_2(m) = 2$  for any m.

For n = 3,  $p_3(0) = 4$ . There are a number of information sets  $\Re$  for which  $p(\Re) = 4$ . A proof and discussion may be found in Meersman [75].

Unfortunately the proof technique used to establish the cases n=2, 3 cannot be used for general n since there are too many sub-cases to investigate.

We now wish to discuss some general properties of the n-evaluation problem.

Recall that  $E_n^k = (e_{ij})$  is the incidence matrix of the information  $\mathfrak N$  and let

(5.1) 
$$M_{r} = \sum_{j=0}^{r} \sum_{i=1}^{k} e_{ij}$$

denote the total number of evaluations  $f, f', ..., f^{(r)}$  at  $z_1, ..., z_k, r = 0, 1, ...$ 

The incidence matrix  $\mathbf{E}_{n}^{k}$  satisfies the <u>Polya conditions</u> if

(5.2) 
$$M_r \ge r+1$$
 for  $r = 0,1,...,n-1$ .

(See Sharma [72].) If  $E_n^k$  satisfies the Polya conditions then  $e_{ij} = 0$  for any i and  $j \ge n$ . This means we do not use

derivatives of order higher than n-1. Note that hermitian  $E_n^k$  satisfies the Polya conditions. Furthermore all known information sets with maximal order of information have  $E_n^k$  which satisfy the Polya conditions.

Let  $j' = j'(E_n^k)$  be a nonnegative integer such that

$$M_r \ge r+1 \text{ for } r = 0,1,...,j' \text{ and } M_{j'+1} < j'+2.$$

Since  $j'+1 \le M_j$ ,  $\le M_{j'+1} \le j'+1$  then  $e_{i,j'+1} = 0$  which means that we do not use the (j'+1) derivative. We shall call such  $j' = j'(E_n^k)$  an index of  $E_n^k$ .  $E_n^k$  satisfies the Polya conditions if and only if its index is equal to n-1.

We introduce the concept of the polynomial order of information  $pol(\mathfrak{N})$  defined by

(5.3) pol(
$$\mathfrak{N}$$
) = 
$$\begin{cases} 0 & \text{if B is empty} \\ \sup B & \text{otherwise} \end{cases}$$

where

$$B = \{q \ge 1 : \forall f \in \mathfrak{J}, \ f(\alpha) = 0, \ \forall \tilde{f} \equiv f \text{ and } \tilde{f} - f \in \Pi_n,$$

$$\lim_{x_1 \to \alpha} \sup_{|x_1 - \alpha|} \frac{|\alpha - \tilde{\alpha}|}{|x_1 - \alpha|^{q - \varepsilon}} = 0, \ \forall \varepsilon > 0\},$$

and  $\Pi_n$  denotes a class of polynomials of degree  $\leq n$ . Compare with the order of information where is not assumed that  $\tilde{f}$ - $f \in \Pi_n$ , see (2.10). Thus  $p(\mathfrak{R}) \leq pol(\mathfrak{R})$ . Similarly let  $pol(n) = \sup_{\mathfrak{R}} p(\mathfrak{R})$ . This gives

$$(5.4) \quad p_n(0) \leq pol(n).$$

We show some properties of pol(n). From Section 4 it follows that pol(n)  $\ge 2^{n-1}$  and pol(n) =  $2^{n-1}$  for hermitian

information. Furthermore it is possible to show that  $pol(n) = 2^{n-1}$  for n = 1,2,3 and that pol(n) is an increasing function of n.

### Lemma 5.1

Let j' be the index of the incidence matrix  $\textbf{E}_{n}^{k}$  of  $\mathfrak{N}.$  Then

$$pol(\mathfrak{N}) \leq pol(j'+1).$$

<u>Proof</u> (Compare with the proof of the Schoenberg Lemma in Schoenberg [66] and Sharma [72], Lemma 1.)

Let  $E_j^k$  denote the first (j'+1) columns of  $E_n^k$ . Assume  $f \in \Pi_{j'+1}$ . Then  $z_i = z_i(x_1; \Re(x_1; f)) = z_i(x_1; \Re_1(x_1; f))$  where  $\Re_1$  is the information based on  $E_j^k$ . Let  $h \in \Pi_{j'+1}$  and

(5.5) 
$$h^{(j)}(z_i) = 0$$
 for  $(i,j) \in e_n^k$  and  $j \leq j'$ .

The total number of homogeneous equations in (5.5) is equal to M<sub>j'</sub> = j'+1 and since we have j'+2 unknowns then there exists a nonzero h satisfying (5.5). Furthermore  $h^{(j)}(x) \equiv 0$  for  $j \geq j'+2$  which means that  $h^{(j)}(z_j) = 0$  for all  $(i,j) \in e_n^k$ . Define  $\tilde{f}(x) = f(x) + h(x)$  we get

(5.6) 
$$\tilde{\alpha} - \alpha = \frac{1}{g'(\alpha)} (1 + o(1))h(\alpha)$$
.

But  $h(\alpha)$  depends only on  $E_j^k$ , and it can be at most of order pol(j'+1). This proves that  $pol(\mathfrak{N}) \leq pol(j'+1)$ .

Since pol(n) is an increasing function of n we immediately have

## Corollary 5.2

A necessary condition for  $\Re$  to have the maximal polynomial order pol(n) is that its incidence matrix  $\mathbb{E}_n^k$  satisfies

the Polya conditions.

We believe that  $pol(n) = 2^{n-1}$ . However to find even a crude upper bound on pol(n) seems to be hard. We give an upper bound on pol(n) under the following conjecture.

### Conjecture 5.3

Let  $\phi_1,\phi_2,\dots,\phi_n$  be any n-point iterations. Then there exists a function  $f\in \Im$  such that

$$(5.7) \quad \lim_{x_1 \to \alpha} \left| \frac{\varphi_{\mathbf{i}}(x_1; \Re(x_1; f)) - \alpha}{e_1^{\text{pol}(n) + \varepsilon}} \right| = +\infty, \forall \varepsilon > 0, \forall i \le n.$$

Assume for simplicity that  $C_i = C_i(f, \varphi_i) = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x_1; \varphi_i)| = \lim_{\substack{x_1 \to \alpha \\ y = 1}} |\varphi_i(x$ 

### <u>Lemma 5.4</u>

If (5.7) holds then pol(n) < n! for  $n \ge 3$ .

### Proof

Let  $E_n^k$  be the incidence matrix of m. Let  $0 \neq h \in \Pi_n$  and  $h^{(j)}(z_i) = 0$  for  $(i,j) \in e_n^k$ . Then

$$h(x;x_1) = a(x_1)(x-h_1)(x-h_2)...(x-h_i)$$

where  $1 \le j \le n$  and  $a(x_1)$  is chosen in order to ensure that  $h(x;x_1)$  tends to an analytic function as  $x_1$  tends to  $\alpha$ . Note that  $h_1 = x_1$  and  $h_i = h_1(z_1,z_2,\ldots,z_k)$  depends on at most (n-1) evaluations. If  $\lim_{x_1 \to \alpha} h_1 = \alpha$  then  $h_1$  can be treated as an iteration. From (5.7) we get

$$|h_i - \alpha| \ge c |e_i|^{pol(n-1)+1-\epsilon}, c > 0,$$

for any  $\epsilon > 0$ . Since it holds for any  $\mathfrak N$  we have

$$pol(n) \le (n-1) pol(n-1) + 1 < n pol(n-1) \le n'$$

The next part of this section deals with a restrictive class of n-point iterations. We use n evaluations per step and we assume that an iteration is exact for a function  $f \in \Pi_{n-1}.$  We shall say that  $\varphi \in \Phi_n$  if  $\varphi(x_1; \Re(x_1; f)) = \alpha$  whenever  $f \in \Pi_{n-1}$  and  $x_1$  is close to  $\alpha$ . Note that all iterations considered in Section 4 belong to  $\Phi_n$ .

Next we shall say that the problem is  $\frac{locally\ well-poised}{n-1}$  for every  $h\in\Pi_{n-1}$  such that

$$h^{(j)}(z_i) = 0$$
 for  $(i,j) \in e_n^k$ 

it follows  $h \equiv 0$  for all  $x_1$  close to  $x_1$ .

Note that Birkhoff interpolation for  $E_n^k$  is well-poised if  $\forall (x_1, x_2, \dots, x_k)$   $h^{(j)}(z_i) = 0$  for  $(i,j) \in e_n^k$  and  $h \in \Pi_{n-1} \Rightarrow h \equiv 0$  (see Sharma [72]). Thus, if Birkhoff interpolation is well-poised than the problem is locally well-poised but not in general vice versa.

# Lemma 5.5

If an iteration  $\phi$  is exact for  $f\in \Pi_{n-1},\; \phi\in \Phi_n,$  then

- (i)  $E_n^k$  satisfies the Polya conditions,
- (ii) the problem is locally well-poised for  $f \in \Pi_{n-1}$ ,
- (iii)  $p(\mathfrak{N}) \leq n(n+1)^{n-1}$ .

## Proof

Suppose that the problem is not locally well-poised for  $f\in\Pi_{n-1}$  . Then there exists a nonzero  $h\in\Pi_{n-1}$  such that

 $h^{(j)}(z_j) = 0$  for  $(i,j) \in e_n^k$ . Define  $\tilde{f}(x) = f(x) + h(x)$ . Since  $\tilde{\tilde{\mathbf{f}}} \in \Pi_{n-1}$  and  $\tilde{\mathbf{f}}(\alpha) \neq 0$  then

$$\alpha = \varphi(\mathbf{x}_1, \mathfrak{N}(\mathbf{x}_1, \mathbf{f})) = \varphi(\mathbf{x}_1, \mathfrak{N}(\mathbf{x}_1, \tilde{\mathbf{f}})) \neq \tilde{\alpha}.$$

This contradicts that  $\mathfrak{g}\in \Phi_{\mathbf{n}}.$  Hence (ii) holds. Let j' be the index of  $E_n^k$ . If j' < n-1 then there exists a nonzero  $h \in \Pi_{j'+1}$  such that  $h^{(j)}(z_i) = 0$  for all  $(i,j) \in e_n^k$ , see the proof of Lemma (5.1). This contradicts that the problem is locally well-poised. Thus, (i) holds.

To prove (iii) it suffices to note that if

$$E_n^k \le \widetilde{E}_{\widetilde{n}}^k$$
 then  $p(E_n^k) \le p(\widetilde{E}_{\widetilde{n}}^k)$ 

for  $n \le \tilde{n}$  where by  $\tilde{E}_n^k = (e_{ij}) \le \tilde{E}_{\tilde{n}}^k = (\tilde{e}_{ij})$  we mean 

$$\tilde{e}_{ij} = 1$$
 for  $i = 1,2,...,k$  and  $j = 0,1,...,n-1$ .

Of course  $E_n^k \le \widetilde{E}_{\widetilde{n}}^k$  and from Theorem 4.1 we get

$$p(\widetilde{E}_{\widetilde{n}}^{k}) \leq n(n+1)^{n-1}$$

which proves (iii).

### FINAL REMARKS 6.

The problem of the maximal order of n-point iterations is connected with Birkhoff interpolation which has been open almost 70 years. The main difficulty is to estimate the difference between the zeros,  $\tilde{\alpha}$ - $\alpha$ , of any two functions with the same information,  $\tilde{f} = \tilde{f}$ . Note that  $\tilde{f}$  can belong to  $\prod_{n=1}^{n}$  for

all f if the problem is well-poised. However up to now we do not know when Birkhoff interpolation is well-poised. There are many reasons to believe that hermitian information (interpolation without gaps) is optimal. However there also exists nonhermitian information with order  $2^{n-1}$ .

For nonhermitian information  $\mathfrak R$  it is hard to find the order  $p(\mathfrak R)$ . We know the order of such information only in a few cases. The first one is a Brent iteration based on  $\mathfrak R=\{f(z_1),f'(z_1),\ldots,f^{(j)}(z_1),f^{(r)}(z_2),f^{(r)}(z_3),\ldots,f^{(r)}(z_k)\}$  for suitable chosen  $z_i$  where  $0< r \leq j+1$  (see Brent [75]). This information uses n=j+k evaluations and has the order  $p(\mathfrak R)=j+2k-1$ , see Meersman [75]. Note that this problem is well-poised. The second example is Abel-Goncarov information given by

$$\mathfrak{N} = \{f(z_1), f'(z_2), \dots, f^{(n-1)}(z_n)\},\$$

see Sharma [72]. Recall that if  $z_i = z_1$  for i = 2, ..., n then we get one-point information which has the order n (even in the multivariate and abstract cases). For Abel-Goncarov information it is possible to prove

$$n \le p(\mathfrak{N}) \le 2n$$

but we do not know whether this upper bound is sharp. Final-ly let us mention lacunary information given by

$$\mathfrak{N} = \{f(z_1), f''(z_1), f(z_2), f''(z_2), \dots, f(z_k), f''(z_k)\}$$

and n = 2k, see Sharma [72]. It is possible to verify that

$$\frac{1}{2} 2^{n/2} \le p(\mathfrak{N}) \le \frac{3}{4} 2^n$$

but the exact value of  $p(\mathfrak{N})$  is unknown.

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